



Please find below the README information of the required contents for code and software submission checklist. We provide step-by-step instructions to run the code for our work entitled **“Dynamic functional connectivity encodes generalizable representations of emotional arousal across individuals and situational contexts”**.

### 1. System requirements

MATLAB\_R2022b, python 3.9.0 with the following packages installed: numpy, pandas, os, sklearn, scipy, warnings. No special non-standard hardware is required.

### 2. Installation guide

Running the scripts requires Jupyter notebook, which can be downloaded from the two websites below. Typical downloading time should be less than 5 minutes.

- 1) <https://www.anaconda.com/download>
- 2) <https://jupyter.org/install>

### 3. Demo

Compiled source code and data: AffectPrediction.zip. Please download the file here: <https://uchicago.box.com/s/pgdyc5pfcy17r73ysy81om739r8nyddv>

Upon downloading the file, we suggest you running the scripts in the order below. Just to note: no change of file path is needed. All the code and data are arranged in relative path.

#### Step 1:

##### Calculating dynamic functional connectivity from 122-ROI-based BOLD time series.

- 1) Code path: `./AffectPrediction/code/a_preprocess`.
- 2) Run `a_slidingFC_Sherlock.m`, `b_slidingFC_FNL.m`, and `c_slidingFC_Merlin.m` in MATLAB. The scripts take 122-ROI-based BOLD time series (`./AffectPrediction/data/brain/{dataset }/a_output/ROIsum.mat`), compute dynamic functional connectivity, and save the FC matrices at (`./AffectPrediction/data/brain/{dataset}/a_output/FC/sliding-dynFeat.mat`).
- 3) This process can take 1-3 hours depending on the computing power of your device. To save you some time, I included these FC files in the folder but feel free to try running the script yourself.

#### Step 2:

##### Within- and Across-dataset predictions on valence and arousal.

- 1) Code path: `./AffectPrediction/code/b_analysis`. These analyses correspond to the following sections in the Results part of the manuscript:
  - *Dynamic functional connectivity encodes arousal within and across datasets*
  - *Connectome-based models of arousal generalized to two more fMRI datasets*
  - *Dynamic functional connectivity does not predict moment-to-moment valence*
- 2) Open Jupyter Notebook and please run the scripts in a-e order. For each script, we suggest you go to Kernel – Restart & Run All.
  - a) `a_Within_Sherlock.ipynb`. Within-dataset predictions of valence and arousal in *Sherlock*, saves results at `./AffectPrediction/results/sherlock/within_prediction` for

later use of across-dataset predictions. This process can take 10-20 minutes depending on the computing power of your device.

- b) *b\_Within\_FNL.ipynb*. similar as a) but in *Friday Night Lights*. This process can take 30-60 minutes depending on the computing power of your device
- c) *c\_Across\_Sher-FNL.ipynb*. Across-dataset predictions of valence and arousal from *Sherlock* to *Friday Night Lights*. This should take less than 5 minutes.
- d) *d\_Across\_FNL-Sher.ipynb*. Across-dataset predictions of valence and arousal from *Friday Night Lights* to *Sherlock*. This should take less than 5 minutes
- e) *e\_overlap-Merlin\_arousal.ipynb*. This script takes results from within-dataset predictions to compute an overlap arousal network (Fig. 4B in the manuscript), then train a CPM on the overlap arousal network by combining data from *Sherlock* and *Friday Night Light* to predict arousal in *Merlin*. This should take less than 5 minutes

### Step 3:

**Applying the arousal model that we identified to your own fMRI data, to generate a moment-to-moment prediction of the arousal fluctuations.**

- a) Parcellate your brain data into 122 ROIs, Yeo 114 for cortical ROIs, and Brainnectome for the 8 subcortical regions (following this order: left amygdala, hippocampus, basal ganglia, and thalamus, right amygdala, hippocampus, basal ganglia, and thalamus).
- b) Calculate dynamic functional connectivity patterns using this code by Hayoung Song: [https://github.com/hyssong/NarrativeEngagement/blob/main/code/preprocess/sliding\\_FC.m](https://github.com/hyssong/NarrativeEngagement/blob/main/code/preprocess/sliding_FC.m) The final output should be: `nsubj*nFC(8376)*nT(nTR-sliding window size)`
- c) Run this code: `./arousal_network/apply_model.ipynb`. This code uses the pre-defined arousal network (`arousal_network.mat`) and the pre-trained SVR (`svr_model.pkl`)

## 4. Instruction to use

For MATLAB scripts: open MATLAB and double click the scripts

For python scripts: Open Anaconda-Navigator and click the “Launch” icon under Jupyter Notebook. This will open a webpage in your browser.

### Additional information:

Link to code in an open source repository (we have it on github):

<https://github.com/jinke828/AffectPrediction>

Please do not hesitate to reach out if you have any questions.

Sincerely,



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